What Is Claimed Is:

- 1 1. A method for using interval techniques within a computer system
- 2 to solve a multi-objective optimization problem, comprising:
- receiving a representation of multiple objective functions $(f_1, ..., f_n)$ at
- 4 the computer system, wherein $(f_1, ..., f_n)$ are scalar functions of a vector
- 5 $\mathbf{x} = (x_1, ..., x_n);$
- 6 receiving a representation of a domain of interest for the multiple objective
- 7 functions;
- 8 storing the representations in a memory within the computer system; and
- 9 performing an interval optimization process to compute guaranteed bounds
- on a Pareto front for the objective functions $(f_1, ..., f_n)$, wherein for each point
- on the Pareto front, an improvement in one objective function cannot be made
- without adversely affecting at least one other objective function;
- wherein performing the interval optimization process involves applying a
- 14 direct-comparison technique between subdomains of the domain of interest to
- eliminate subdomains that are certainly dominated by other subdomains.
- 1 2. The method of claim 1, wherein performing the interval
- 2 optimization process involves applying a gradient technique to eliminate
- 3 subdomains that do not contain a local Pareto optimum.
- 1 3. The method of claim 2, wherein a subdomain $[x]_i$ is eliminated by
- 2 the gradient technique if an intersection of certainly negative gradient regions C_j
- 3 for each objective function f_j is non-empty, $\bigcap_{j=1}^n \mathbf{C}_j([\mathbf{x}]_j) \neq \emptyset$;

1	wherein the certainly	v nagativa gradiani	tracion C for	abjective function f	:.
4	wherein the certaining	y negative gradieni	\mathbf{c}_i region \mathbf{c}_i for	objective function f_i	18

- 5 the intersection of $N_j([x]_i)$ (the negative gradient region associated with the
- 6 minimum angle θ_j of the gradient of f_j over the subdomain $[\mathbf{x}]_i$ and $\overline{\mathbf{N}_j}([\mathbf{x}]_i)$ (the
- 7 negative gradient region associated with the maximum angle $\overline{\theta_j}$ of the gradient of
- 8 f_j over the subdomain $[\mathbf{x}]_i$).
- 1 4. The method of claim 3, wherein the method further comprises
- 2 iteratively:
- 3 bisecting remaining subdomains that have not been eliminated by the
- 4 gradient technique; and
- 5 applying the gradient technique to eliminate bisected subdomains that do
- 6 not contain a local Pareto optimum.
- 1 5. The method of claim 4, wherein bisecting a subdomain involves
- 2 bisecting the subdomain in the direction that has the largest width of partial
- derivatives of all objective functions $(f_1, ..., f_n)$ over the subdomain.
- 1 6. The method of claim 4, wherein the direct-comparison technique is
- 2 applied once for every n iterations of the gradient technique.
- The method of claim 6, wherein the iterations continue until either
- 2 a predetermined maximum number of iterations are performed, or the largest area-
- 3 of any subdomain is below a predetermined value.
 - 8. The method of claim 1,

1

- wherein a subdomain **U** certainly dominates a subdomain **V** if every point
- 3 $\mathbf{u} \in \mathbf{U}$ dominates every point $\mathbf{v} \in \mathbf{V}$; and
- 4 wherein a point **u** dominates a point **v** under minimization if,
- 5 $u_i \le v_i, i = 1, ..., n$, and
- 6 $u_i < v_i$ for some $i \in \{1, ..., n\}$.
- 9. A computer-readable storage medium storing instructions that
- 2 when executed by a computer cause the computer to perform a method for using
- 3 interval techniques within a computer system to solve a multi-objective
- 4 optimization problem, the method comprising:
- receiving a representation of multiple objective functions $(f_1, ..., f_n)$ at
- 6 the computer system, wherein $(f_1, ..., f_n)$ are scalar functions of a vector
- 7 $\mathbf{x} = (x_1, ..., x_n);$
- 8 receiving a representation of a domain of interest for the multiple objective
- 9 functions;
- storing the representations in a memory within the computer system; and
- performing an interval optimization process to compute guaranteed bounds
- on a Pareto front for the objective functions $(f_1, ..., f_n)$, wherein for each point
- on the Pareto front, an improvement in one objective function cannot be made
- without adversely affecting at least one other objective function;
- wherein performing the interval optimization process involves applying a
- direct-comparison technique between subdomains of the domain of interest to
- eliminate subdomains that are certainly dominated by other subdomains.
 - 1 10. The computer-readable storage medium of claim 9, wherein
 - 2 performing the interval optimization process involves applying a gradient
 - 3 technique to eliminate subdomains that do not contain a local Pareto optimum.

- 1 11. The computer-readable storage medium of claim 10, wherein a
- subdomain $[x]_i$ is eliminated by the gradient technique if an intersection of
- 3 certainly negative gradient regions C_i for each objective function f_i is non-empty,
- $4 \qquad \bigcap_{j=1}^{n} \mathbf{C}_{j}([\mathbf{x}]_{j}) \neq \emptyset;$
- 5 wherein the certainly negative gradient region C_j for objective function f_j is
- 6 the intersection of $N_j([x]_i)$ (the negative gradient region associated with the
- 7 minimum angle $\underline{\theta_j}$ of the gradient of f_j over the subdomain $[\mathbf{x}]_i$) and $\overline{\mathbf{N}_j}([\mathbf{x}]_i)$ (the
- 8 negative gradient region associated with the maximum angle $\overline{\theta_j}$ of the gradient of
- 9 f_i over the subdomain $[\mathbf{x}]_i$).
- 1 12. The computer-readable storage medium of claim 11, wherein the
- 2 method further comprises iteratively:
- 3 bisecting remaining subdomains that have not been eliminated by the
- 4 gradient technique; and
- 5 applying the gradient technique to eliminate bisected subdomains that do
- 6 not contain a local Pareto optimum.
- 1 13. The computer-readable storage medium of claim 12, wherein
- 2 bisecting a subdomain involves bisecting the subdomain in the direction that has
- 3 the largest width of partial derivatives of all objective functions $(f_1, ..., f_n)$ over
- 4 the subdomain.

- 1 14. The computer-readable storage medium of claim 12, wherein the
- 2 direct-comparison technique is applied once for every n iterations of the gradient
- 3 technique.
- 1 15. The computer-readable storage medium of claim 14, wherein the
- 2 iterations continue until either a predetermined maximum number of iterations are
- 3 performed, or the largest area of any subdomain is below a predetermined value.
- 1 16. The computer-readable storage medium of claim 9,
- wherein a subdomain U certainly dominates a subdomain V if every point
- 3 $\mathbf{u} \in \mathbf{U}$ dominates every point $\mathbf{v} \in \mathbf{V}$; and
- 4 wherein a point **u** dominates a point **v** under minimization if,
- 5 $u_i \le v_i, i = 1, ..., n$, and
- 6 $u_i < v_i \text{ for some } i \in \{1, ..., n\}.$
- 1 17. An apparatus that uses interval techniques to solve a multi-
- 2 objective optimization problem, comprising:
- a receiving mechanism configured to receive a representation of multiple
- 4 objective functions $(f_1, ..., f_n)$, wherein $(f_1, ..., f_n)$ are scalar functions of a
- 5 vector $\mathbf{x} = (x_1, ..., x_n);$
- 6 wherein the receiving mechanism is configured to receive a representation
- 7 of a domain of interest for the multiple objective functions;
- 8 a memory configured to store the representations; and
- 9 an interval optimizer configured to performing an interval optimization
- 10 process to compute guaranteed bounds on a Pareto front for the objective
- functions $(f_1, ..., f_n)$, wherein for each point on the Pareto front, an

- 12 improvement in one objective function cannot be made without adversely
- affecting at least one other objective function;
- wherein the interval optimizer is configured to apply a direct-comparison
- 15 technique between subdomains of the domain of interest to eliminate subdomains
- that are certainly dominated by other subdomains.
- 1 18. The apparatus of claim 17, wherein the interval optimizer is
- 2 configured to apply a gradient technique to eliminate subdomains that do not
- 3 contain a local Pareto optimum.
- 1 19. The apparatus of claim 18, wherein a subdomain $[x]_i$ is eliminated
- 2 by the gradient technique if an intersection of certainly negative gradient regions
- 3 \mathbf{C}_j for each objective function f_j is non-empty, $\bigcap_{j=1}^n \mathbf{C}_j([\mathbf{x}]_j) \neq \emptyset$;
- 4 wherein the certainly negative gradient region C_i for objective function f_i is
- 5 the intersection of $\mathbf{N}_{j}([\mathbf{x}]_{i})$ (the negative gradient region associated with the
- 6 minimum angle $\underline{\theta_j}$ of the gradient of f_j over the subdomain $[\mathbf{x}]_i$ and $\overline{\mathbf{N}_j}([\mathbf{x}]_i)$ (the
- 7 negative gradient region associated with the maximum angle $\overline{\theta_j}$ of the gradient of
- 8 f_j over the subdomain $[\mathbf{x}]_i$).
- 1 20. The apparatus of claim 19, wherein the interval optimizer is
- 2 configured to iteratively:
- 3 bisect remaining subdomains that have not been eliminated by the gradient
- 4 technique; and to
- 5 apply the gradient technique to eliminate bisected subdomains that do not
- 6 contain a local Pareto optimum.

- 1 21. The apparatus of claim 20, wherein bisecting a subdomain involves
- 2 bisecting the subdomain in the direction that has the largest width of partial
- derivatives of all objective functions $(f_1, ..., f_n)$ over the subdomain.
- 1 22. The apparatus of claim 20, wherein the direct-comparison
- 2 technique is applied once for every *n* iterations of the gradient technique.
- 1 23. The apparatus of claim 22, wherein the iterations continue until
- 2 either a predetermined maximum number of iterations are performed, or the
- 3 largest area of any subdomain is below a predetermined value.
- 1 24. The apparatus of claim 17,
- wherein a subdomain U certainly dominates a subdomain V if every point
- 3 $\mathbf{u} \in \mathbf{U}$ dominates every point $\mathbf{v} \in \mathbf{V}$; and
- 4 wherein a point **u** dominates a point **v** under minimization if,
- 5 $u_i \le v_i, i = 1, ..., n$, and
- 6 $u_i < v_i \text{ for some } i \in \{1, ..., n\}.$